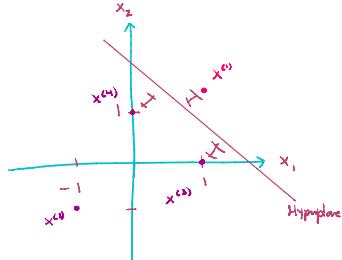


Dataset:

$$\begin{aligned}x^{(1)} &= \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad x^{(2)} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}, \quad x^{(3)} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad x^{(4)} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \\ \uparrow & \quad \uparrow & \quad \uparrow & \quad \uparrow \\ \text{Positive} & \quad \text{Negative} & \quad \text{negative} & \quad \text{negative}\end{aligned}$$

or: $y^{(1)} = 1, \quad y^{(2)} = y^{(3)} = y^{(4)} = -1$

1.



2. Support Vectors:

$$x^{(1)}, x^{(2)}, x^{(3)}$$

3. Opt. Problem: $\min_{\vec{w}, b} \frac{1}{2} \|\vec{w}\|^2 \quad \text{s.t. } \forall j \quad (\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} \geq 1$

Write down the lagrangian:

$$\mathcal{L}(\vec{w}, \lambda) = f_0(\vec{x}) + \sum_{i=1}^m \lambda_i f_i(\vec{x})$$

$$f_0(\vec{x}) = \frac{1}{2} \|\vec{w}\|^2$$

$$f_j(\vec{x}) = j (\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1 \geq 0$$

$$\therefore \mathcal{L}(\vec{w}, b, \vec{\alpha}) = \frac{1}{2} \|\vec{w}\|^2 - \sum_{j=1}^m \alpha_j ((\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)} - 1)$$

4. $\min_{\vec{w}, b} \max_{\vec{\alpha}} \mathcal{L}(\vec{w}, b, \vec{\alpha})$

with Slater's Condition

$$\max_{\vec{\alpha}} \min_{\vec{w}, b} \mathcal{L}(\vec{w}, b, \vec{\alpha})$$

Solve min problem:

$$\max_{\vec{w}} \min_{b} \frac{1}{2} \|\vec{w}\|^2 - \sum_{j=1}^m \alpha_j ((\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)})$$

$$= \max_{\vec{w}} \min_{b} \frac{1}{2} \|\vec{w}\|^2 - \sum_{j=1}^m \alpha_j (\vec{w} \cdot \vec{x}^{(j)} y^{(j)} + \alpha_j \cdot b \cdot y^{(j)}) - \alpha_j$$

$$\hookrightarrow \frac{1}{2} \|\vec{w}\|^2 = \vec{w}$$

$$\frac{\partial L}{\partial \vec{w}} = \vec{w} - \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)}$$

$$0 = \vec{w} - \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)}$$

$$\vec{w} = \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)}$$

$$\frac{\partial L}{\partial b} = \sum_{j=1}^m \alpha_j \cdot y^{(j)}$$

$$0 = \sum_{j=1}^m \alpha_j y^{(j)}$$

$$\therefore \vec{A} = \vec{w}, B_j = \alpha_j, C_j = y^{(j)}, D_j = \vec{x}^{(j)}$$

$$E_j = \alpha_j, F_j = y^{(j)}$$

5. Simplify:

$$\begin{aligned} L(\vec{w}, b, \alpha) &= \frac{1}{2} \|\vec{w}\|^2 - \sum_{j=1}^m \alpha_j ((\vec{w} \cdot \vec{x}^{(j)} + b) y^{(j)}) \\ &= \frac{1}{2} \vec{w}^T \vec{w} + \sum_{j=1}^m \alpha_j - \underbrace{\sum_{j=1}^m \alpha_j \vec{w}^T \vec{x}^{(j)} y^{(j)}}_{\text{from above, } \sum_{j=1}^m \alpha_j y^{(j)} = 0} - \underbrace{\sum_{j=1}^m \alpha_j y^{(j)} b}_{\therefore \text{whole thing} = 0} \end{aligned}$$

$$= \frac{1}{2} \vec{w}^T \vec{w} + \sum_{j=1}^m \alpha_j - \sum_{j=1}^m \alpha_j \vec{w}^T \vec{x}^{(j)} y^{(j)}$$

$$= \sum_{j=1}^m \alpha_j + \vec{w}^T \left(\frac{1}{2} \vec{w} - \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)} \right)$$

From above

$$= \sum_{j=1}^m \alpha_j + \vec{w}^T \left(\frac{1}{2} \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)} - \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)} \right)$$

From above

$$= \sum_{j=1}^m \alpha_j + w^T \left(\frac{1}{2} \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)} - \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)} \right)$$

$$= \sum_{j=1}^m \alpha_j + w^T \left(-\frac{1}{2} \sum_{j=1}^m \alpha_j \vec{x}^{(j)} y^{(j)} \right)$$

$$w = \sum_{j=1}^m \alpha_j y^{(j)} \vec{x}^{(j)} \rightarrow \text{Vector}$$

$$w^T = \sum_{k=1}^n \alpha_k y^{(k)} \vec{x}^{T(k)}$$

$$= \sum_{j=1}^m \alpha_j - \frac{1}{2} \sum_{j=1}^m \sum_{h=1}^n \alpha_j \alpha_h y^{(j)} y^{(h)} \vec{x}^{T(h)} \vec{x}^{(j)}$$

$$\therefore \max_{\vec{w}} \min_{\vec{w}, b} d(\vec{w}, b, \vec{a}) = \max_{\vec{w}} \sum_{j=1}^m \alpha_j - \frac{1}{2} \sum_{j=1}^m \sum_{h=1}^n \alpha_j \alpha_h y^{(j)} y^{(h)} (\vec{x}^{(h)})^T \vec{x}^{(j)}$$

$$A_j = \alpha_j, B_j = \alpha_j, C_h = \alpha_h, D_j = y^{(j)}, E_h = y^{(h)}$$

$$\vec{F} = (\vec{x}^{(h)})^T \quad \vec{G} = \vec{x}^{(j)}$$

6: From observing the dataset that $x^{(2)} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$

is not a support vector since its not close
to the decision boundary. Therefore $\alpha^{(2)} = 0$

7.

From Question 5.

$$O = \sum_{j=1}^m \alpha_j y^{(j)}$$

$$O = \alpha_1 y^1 + \alpha_2 y^2 + \alpha_3 y^3 + \alpha_4 y^4$$

$$O = \alpha_1(1) + O(-1) + \alpha_3(-1) + \alpha_4(-1)$$

$$0 = \alpha_1 - \alpha_3 - \alpha_4$$

$$\alpha_1 = \alpha_3 + \alpha_4$$

8. Solve Dual Problem

$$\max_{\vec{\alpha}} \min_{\vec{w}, b} L(\vec{w}, b, \vec{\alpha}) = \max_{\vec{\alpha}} \sum_{j=1}^m \alpha_j - \frac{1}{2} \sum_{j=1}^m \sum_{h=1}^n \alpha_j \alpha_h y^{(j)} y^{(h)} (\vec{x}^{(h)})^T \vec{x}^{(j)}$$

$$\text{s.t. } \sum_{j=1}^m \alpha_j y_j = 0 \quad \& \quad \alpha_j \geq 0$$

$$\begin{aligned}
 \alpha_1 &= \alpha_3 + \alpha_4 \\
 &= \max_{\vec{\alpha}} \sum_{j=1}^m \alpha_j - \frac{1}{2} \sum_{j=1}^m \sum_{h=1}^n \alpha_j \alpha_h y^{(j)} y^{(h)} \langle \vec{x}^{(h)}, \vec{x}^{(j)} \rangle \\
 &= \max_{\vec{\alpha}} (\alpha_1 + \alpha_3 + \alpha_4) - \frac{1}{2} \left[(\alpha_1 \alpha_1 (1)(1) [1 1] \begin{bmatrix} 1 \\ 1 \end{bmatrix}) + (\alpha_1 \alpha_3 (1)(-1) [1 1] \begin{bmatrix} 1 \\ 0 \end{bmatrix}) + (\alpha_1 \alpha_4 (1)(-1) [1 1] \begin{bmatrix} 0 \\ 1 \end{bmatrix}) \right. \\
 &\quad \left. + (\alpha_3 \alpha_1 (-1)(1) [1 0] \begin{bmatrix} 1 \\ 1 \end{bmatrix}) + (\alpha_3 \alpha_3 (-1)(-1) [1 0] \begin{bmatrix} 1 \\ 0 \end{bmatrix}) + (\alpha_3 \alpha_4 \cancel{(-1)(-1)} [1 0] \begin{bmatrix} 0 \\ 1 \end{bmatrix}) \right] = 0 \\
 &\quad + (\alpha_4 \alpha_1 (-1)(1) [0 1] \begin{bmatrix} 1 \\ 1 \end{bmatrix}) + (\alpha_4 \alpha_3 \cancel{(-1)(-1)} [0 1] \begin{bmatrix} 1 \\ 0 \end{bmatrix}) + (\alpha_4 \alpha_4 (-1)(-1) [0 1] \begin{bmatrix} 0 \\ 1 \end{bmatrix}) \\
 &= \max_{\vec{\alpha}} (\alpha_1 + \alpha_3 + \alpha_4) - \frac{1}{2} \left[\cancel{\alpha_1^2} - \cancel{\alpha_1 \alpha_3} - \cancel{\alpha_1 \alpha_4} - \cancel{\alpha_3 \alpha_1} + \cancel{\alpha_3^2} + \cancel{\alpha_3 \alpha_4} - \cancel{\alpha_4 \alpha_1} + \cancel{\alpha_4 \alpha_3} + \cancel{\alpha_4^2} \right] \\
 &= \max_{\vec{\alpha}} (\alpha_1 + \alpha_3 + \alpha_4 - \alpha_1^2 + \alpha_1 \alpha_3 + \alpha_1 \alpha_4 - \frac{\alpha_3^2}{2} - \frac{\alpha_4^2}{2}) \\
 &= \max_{\vec{\alpha}} \left(-\alpha_1^2 - \frac{\alpha_3^2}{2} - \frac{\alpha_4^2}{2} + \alpha_1 \alpha_3 + \alpha_1 \alpha_4 + \alpha_1 + \alpha_3 + \alpha_4 \right)
 \end{aligned}$$

Use $\alpha_1 = \alpha_3 + \alpha_4$ to eliminate a variable

$$\begin{aligned}
 &= \max_{\vec{\alpha}} \left(-(\alpha_3 + \alpha_4)^2 - \frac{\alpha_3^2}{2} - \frac{\alpha_4^2}{2} + (\alpha_3 + \alpha_4) \alpha_3 + (\alpha_3 + \alpha_4) \alpha_4 + (\alpha_3 + \alpha_4) + \alpha_3 + \alpha_4 \right) \\
 &= \max_{\vec{\alpha}} \left(-(\alpha_3^2 + 2\alpha_3 \alpha_4 + \alpha_4^2) - \frac{\alpha_3^2}{2} - \frac{\alpha_4^2}{2} + \alpha_3^2 + \alpha_4 \alpha_3 + \alpha_3 \alpha_4 + \alpha_4^2 + 2\alpha_3 + 2\alpha_4 \right) \\
 &= \max_{\vec{\alpha}} \left(-\cancel{\alpha_3^2} - \cancel{2\alpha_3 \alpha_4} - \cancel{\alpha_4^2} - \frac{\alpha_3^2}{2} - \frac{\alpha_4^2}{2} + \cancel{\alpha_3^2} + \cancel{\alpha_3 \alpha_4} + \cancel{\alpha_3 \alpha_4} + \cancel{\alpha_4^2} + 2\alpha_3 + 2\alpha_4 \right)
 \end{aligned}$$

$$\begin{aligned}
 &= \max_{\vec{\alpha}} \left(-\cancel{\alpha_3^2} - 2\cancel{\alpha_3\alpha_4} - \cancel{\alpha_4^2} - \frac{\alpha_3^2}{2} - \frac{\alpha_4^2}{2} + \cancel{\alpha_3^2} + \cancel{\alpha_3\alpha_4} + \cancel{\alpha_3\alpha_4} + \cancel{\alpha_4^2} + 2\alpha_3 + 2\alpha_4 \right) \\
 &= \max_{\vec{\alpha}} \left(-\frac{\alpha_3^2}{2} - \frac{\alpha_4^2}{2} + 2\alpha_3 + 2\alpha_4 \right) \\
 \frac{\partial L}{\partial \vec{\alpha}} &= \begin{bmatrix} -\alpha_3 + 2 & -\alpha_4 + 2 \end{bmatrix}
 \end{aligned}$$

$$0 = -\alpha_3 + 2 \quad \quad \quad 0 = -\alpha_4 + 2$$

$$\alpha_3 = 2 \quad \quad \quad \alpha_4 = 2$$

$$\alpha_1 = \alpha_3 + \alpha_4 = 2 + 2$$

$$\alpha_1 = 4$$

$$\vec{\alpha} = [4 \ 0 \ 2 \ 2]$$

9. Solve for \vec{w}

$$\begin{aligned}
 \vec{w} &= \sum_{j=1}^m \alpha_j \vec{x}^{(j)} \vec{y}^{(j)} \\
 \vec{w} &= 4 \begin{bmatrix} 1 \\ 1 \end{bmatrix} (1) + 0 \begin{bmatrix} -1 \\ -1 \end{bmatrix} (-1) + 2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} (-1) + 2 \begin{bmatrix} 0 \\ 1 \end{bmatrix} (-1) \\
 &= \begin{bmatrix} 4+0-2+0 \\ 4+0+0-2 \end{bmatrix} \\
 &= \begin{bmatrix} 2 \\ 2 \end{bmatrix}
 \end{aligned}$$

10. Find b using constraints in eqn 1:

$$(w^T \vec{x}^{(j)} + b) = y^{(j)}$$

$$b = y^{(j)} - w^T \vec{x}^{(j)}$$

For $j=1$

$$\begin{aligned} b &= 1 - [2 \ 2] \begin{bmatrix} 1 \\ 1 \end{bmatrix} \\ &= 1 - 4 \\ &= -3 \end{aligned}$$

For $j=3$

$$\begin{aligned} b &= -1 - [2 \ 2] \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ &= -1 - 2 \\ &= -3 \end{aligned}$$

For $j=4$

$$\begin{aligned} b &= -1 - [2 \ 2] \begin{bmatrix} 0 \\ 1 \end{bmatrix} \\ &= -3 \end{aligned}$$

$$\therefore b = -3$$

II. Hyperplane: $y = \vec{w}^T \vec{x} + b$

$$y = [2 \ 2]^T \vec{x} - 3$$

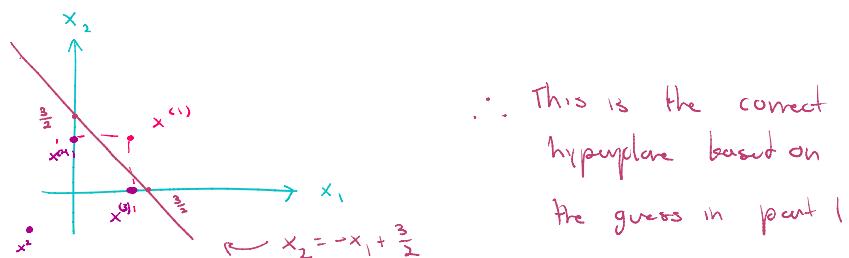
$$y = 2x_1 + 2x_2 - 3$$

Find relationship between x_2 & x_1

$$0 = 2x_1 + 2x_2 - 3$$

$$-2x_2 = 2x_1 - 3$$

$$x_2 = -x_1 + \frac{3}{2}$$



\therefore This is the correct hyperplane based on the guess in point 1